

Custom plotting and related functions for compositional data analysis

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pkde2d --- point density estimation

The **pkde2d()** function is a minor adaptation of the **kde2d()** function found in the MASS package (Venables and Ripley, 2002). Its purpose is to estimate the density of a single datum following 2-dimensional kernel density estimation. It permits the construction of density contours containing a specified percentage of the specimens in a given sample.

Function arguments:

- *x* = x coordinate of data
- *y* = y coordinate of data
- *h* = vector of bandwidths for x and y
- *n* = number of grid points
- *lims* = limits of the rectangle covered by the grid

Function values:

- *x* = the x coordinates of the input data
- *y* = the y coordinates of the input data
- *z* = *n* x *n* matrix of estimated density

See **kde2d()** in package MASS (Venables and Ripley, 2002)

```
pkde2d<-function (x, y, h, n = 25, lims = c(range(x), range(y)))
{
  nx <- length(x)
  if (length(y) != nx)
    stop("data vectors must be the same length")
  if (any(!is.finite(x)) || any(!is.finite(y)))
    stop("missing or infinite values in the data are not allowed")
}
```

```

if (any(!is.finite(lims)))
  stop("only finite values are allowed in 'lims'")
n <- rep(n, length.out = 2L)
h <- if (missing(h))
  c(bandwidth.nrd(x), bandwidth.nrd(y))
else rep(h, length.out = 2L)
h <- h/4
ax <- outer(x, x, "-")/h[1L]
ay <- outer(y, y, "-")/h[2L]
z <- tcrossprod(matrix(dnorm(ax), , nx), matrix(dnorm(ay),
, nx))/(nx * h[1L] * h[2L])

list(x = x, y = y, z = z)
}

```

***geomean()*---calculation of th geometric mean for a given specimen**

The **geomean()** function calculates the geometric mean for a specimen or set of specimens. It is essential for performing centred-log ratio transformations of a dataset.

Function arguments:

- *x = a vector of numeric values*

```

geomean<-function(x){
  log_x<-log(x)
  gm<-exp(mean(log_x))
  return(gm)
}

```

***pcaplot()*---standard PCA plot**

This function generates a PCA plot of standardized data with loadings and scores normalized to the eigenvalues.

Function arguments:

- *x = a matrix or dataframe of numeric values*
- *xx = PC axis to plot on the abscissa (1, 2, 3, ...)*
- *yy = PC axis to plot on the ordinate (1, 2, 3, ...)*
- *group = a vector for defining associated specimens (members of the same group will be assigned the same colour)*

- *label* = a vector of names to plot to the upper right of each point

Function Values:

- *clusters* = data clustered by grouping variable
- *pc* = results of the PCA by **prcomp()**
- *scores* = a dataframe of specimen scores arising from the PCA
- *norm.scores* = a dataframe of eigenvalue normalized scores
- *fc* = a vector of field colours assigned to each group

```

pcaplot<-function(x,xx,yy,group,label)
{
  pca<-prcomp(x,retx=T,scale=T)

  scores<-array(data=NA,dim=c(nrow(pca$x),ncol(pca$x)),
                dimnames=list(rownames(pca$x),colnames(pca$x)))
  for (l in 1:ncol(pca$x)){
    scores[,l]<-pca$x[,l]/pca$sdev[l]
  }

  loadings<-array(data=NA,dim=c(nrow(pca$rotation),
                                ncol(pca$rotation)),
                  dimnames=list(rownames(pca$rotation),
                                colnames(pca$rotation)))

  for (m in 1:ncol(pca$rotation)){
    loadings[,m]<-pca$rotation[,m]*pca$sdev[m]
  }
  percx<-round(pca$sdev[xx]^2/sum(pca$sdev^2)*100,2)
  percy<-round(pca$sdev[yy]^2/sum(pca$sdev^2)*100,2)

  clusters<-list()
  for(i in 1:nlevels(group)){
    clusters[[i]]<-scores[group==levels(group)[i],]
  }

  fc<-rainbow(length(clusters),s=1,v=1,alpha=0.3)
  plot(scores[,xx],scores[,yy],
        pch=NA,
        xlim=c(min(loadings[,xx],scores[,xx])-0.5,
                max(loadings[,xx],scores[,xx])+0.5),
        ylim=c(min(loadings[,yy],scores[,yy])-0.5,

```

```

        max(loadings[,yy],scores[,yy])+0.5),
        xlab=paste(colnames(scores)[xx]," (",percx,"%)",sep=""),
        ylab=paste(colnames(scores)[yy]," (",percy,"%)",sep="")
    )

for(g in 1:length(clusters)){
  if(is.null(nrow(clusters[[g]]))==F){
    points(clusters[[g]][,xx],clusters[[g]][,yy],
           pch=21,
           cex=2,
           col=NA,
           bg=fc[g])
  }
  else{
    points(clusters[[g]][xx],clusters[[g]][yy],
           pch=21,
           cex=2,
           col=NA,
           bg=fc[g])
  }
}
arrows(0,0,loadings[,xx],loadings[,yy],
       length=0.1,
       angle=15
)
text(loadings[,xx]*1.1,loadings[,yy]*1.1,rownames(loadings),
      cex=0.8,
      col="darkblue",
      font=2
)
text(scores[,xx]+0.1,scores[,yy]+0.1,label,cex=0.6,adj=0.25)

list(clusters=clusters, pc=pca, scores=pca$x, norm.scores=scores, fc=fc)
}

```

pcaplot.clr() --- unstandardized PCA plot

This function generates a PCA plot of unstandardized data with loadings and scores normalized to the eigenvalues. It is designed for handling clr-transformed data, though loading vectors may require alternative scaling.

Function arguments:

- $x = \text{data}$
- $xx = \text{PC axis to plot on the abscissa (1, 2, 3, \dots)}$
- $yy = \text{PC axis to plot on the ordinate (1, 2, 3, \dots)}$
- $\text{group} = \text{a vector for defining associated specimens (members of the same group will be assigned the same colour)}$
- $\text{label} = \text{a vector of names to plot to the upper right of each point}$

Function Values:

- $\text{clusters} = \text{data clustered by grouping variable}$
- $\text{pc} = \text{results of the PCA by } \mathbf{prcomp()}$
- $\text{scores} = \text{a dataframe of specimen scores arising from the PCA}$
- $\text{norm.scores} = \text{a dataframe of eigenvalue normalized scores}$
- $\text{fc} = \text{a vector of field colours assigned to each group}$

```

pcaplot.clr<-function(x,xx,yy,group,label)
{
  pca<-prcomp(x,retx=T,scale=F)

  scores<-array(data=NA,dim=c(nrow(pca$x),ncol(pca$x)),
                dimnames=list(rownames(pca$x),colnames(pca$x)))
  for (l in 1:ncol(pca$x)){
    scores[,l]<-pca$x[,l]/pca$sdev[l]
  }

  loadings<-array(data=NA,dim=c(nrow(pca$rotation),
                                ncol(pca$rotation)),
                 dimnames=list(rownames(pca$rotation),
                                colnames(pca$rotation)))

  for (m in 1:ncol(pca$rotation)){
    loadings[,m]<-pca$rotation[,m]*pca$sdev[m]
  }
  percx<-round(pca$sdev[xx]^2/sum(pca$sdev^2)*100,2)
  percy<-round(pca$sdev[yy]^2/sum(pca$sdev^2)*100,2)

  clusters<-list()
  for(i in 1:nlevels(group)){
    clusters[[i]]<-scores[group==levels(group)[i],]
  }
}

```

```

fc<-rainbow(length(clusters),s=1,v=1,alpha=0.3)
plot(scores[,xx],scores[,yy],
      pch=NA,
      xlim=c(min(loadings[,xx],scores[,xx])-0.5,
              max(loadings[,xx],scores[,xx])+0.5),
      ylim=c(min(loadings[,yy],scores[,yy])-0.5,
              max(loadings[,yy],scores[,yy])+0.5),
      xlab=paste(colnames(scores)[xx]," (",percx,"%)",sep=""),
      ylab=paste(colnames(scores)[yy]," (",percy,"%)",sep="")
)

for(g in 1:length(clusters)){
  if(is.null(nrow(clusters[[g]]))==F){
    points(clusters[[g]][,xx],clusters[[g]][,yy],
           pch=21,
           cex=2,
           col=NA,
           bg=fc[g])}
  else{
    points(clusters[[g]][xx],clusters[[g]][yy],
           pch=21,
           cex=2,
           col=NA,
           bg=fc[g])
  }
}
arrows(0,0,loadings[,xx],loadings[,yy],
       length=0.1,
       angle=15
)
text(loadings[,xx]*1.1,loadings[,yy]*1.1,rownames(loadings),
      cex=0.8,
      col="darkblue",
      font=2
)
text(scores[,xx]+0.1,scores[,yy]+0.1,label,cex=0.6,adj=0.25)

list(clusters=clusters, pc=pca, scores=pca$x, norm.scores=scores, fc=fc)
}

```

`pcaplot.kde()`---standard PCA plot with KDEs

This function generates a PCA plot of standardized data with loadings and scores normalized to the eigenvalues. It also generates 100% KDE contours around defined training sets.

Function arguments:

- `x` = data
- `xx` = PC axis to plot on the abscissa (1, 2, 3, ...)
- `yy` = PC axis to plot on the ordinate (1, 2, 3, ...)
- `ts` = a vector for defining specimen groups or training sets (members of the same group will be assigned the same colour)
- `fc` = a vector of field colour choices points and 2D KDE contours for each training set

Function Values:

- `groups` = data clustered by grouping variable
- `pc` = results of the PCA by `prcomp()`

```
pcaplot.kde<-function(x,xx,yy,ts,fc)
{
  pca<-prcomp(x,retx=T,scale=T)
  scores<-array(data=NA,
                dim=c(nrow(pca$x),ncol(pca$x)),
                dimnames=list(rownames(pca$x),colnames(pca$x))
                )
  for (l in 1:ncol(pca$x)){
    scores[,l]<-pca$x[,l]/pca$sdev[l]
  }
  loadings<-array(data=NA,
                  dim=c(nrow(pca$rotation),ncol(pca$rotation)),
                  dimnames=list(rownames(pca$rotation),colnames(pca$rotation))
                  )
  for (m in 1:ncol(pca$rotation)){
    loadings[,m]<-pca$rotation[,m]*pca$sdev[m]
  }
  percx<-round(pca$sdev[xx]^2/sum(pca$sdev^2)*100,2)
  percy<-round(pca$sdev[yy]^2/sum(pca$sdev^2)*100,2)

  groups<-list()
  for(i in 1:nlevels(ts)){
    groups[[i]]<-subset(scores,ts==levels(ts)[i])
  }
}
```

```

}
widths<-array(data=NA,c(length(groups),2))
for(j in 1:length(groups)){
  ifelse(nrow(groups[[j]])<4,
    widths[j,1]<-NA,
    widths[j,1]<-width.SJ(groups[[j]][,xx],method="dpi")
  )
  ifelse(nrow(groups[[j]])<4,
    widths[j,2]<-NA,
    widths[j,2]<-width.SJ(groups[[j]][,yy],method="dpi")
  )
}
kde<-list()
for(k in 1:length(groups)){
  if(is.na(widths[k,1])){
    kde[[k]]<-NA
  }
  else {
    kde[[k]]<-kde2d(as.vector(groups[[k]][,xx]),as.vector(groups[[k]][,yy])
,
    h=c(widths[k,1],widths[k,2]),
    n=c(1000,1000),
    lims=c(range(scores[,xx]-5,scores[,xx]+5),
    range(scores[,yy]-5,scores[,yy]+5))
  )
}
}

pkde<-list()
for(p in 1:length(groups)){
  if(is.na(widths[p,1])){
    pkde[[p]]<-NA
  }
  else {
    pkde[[p]]<-pkde2d(groups[[p]][,xx],groups[[p]][,yy],
    h=c(widths[p,1],widths[p,2]),
    n=c(1000,1000),
    lims=c(range(scores[,xx]-5,scores[,xx]+5),
    range(scores[,yy]-5,scores[,yy]+5))
  )
}
}

```

```

qe<-seq(0,1,len=10001)

quant<-list()
n_groups<-NULL
for (p in 1:length(groups)){
  if(is.na(widths[p,1])){
    quant[[p]]<-NA
  }
  else {
    n_groups[p]<-nrow(groups[[p]])
    ptdens<-NULL
    for(k in 1:n_groups[p]){
      ptdens<-rbind(ptdens,pkde[[p]]$z[k,k])
    }
    quant[[p]]<-quantile(ptdens,probs=qe)
  }
}

plot(scores[,xx],scores[,yy],
      pch=NA,
      xlim=c(min(scores[,xx])-0.5,max(scores[,xx])+0.5),
      ylim=c(min(scores[,yy])-0.5,max(scores[,yy])+0.5),
      xlab=paste(colnames(scores)[xx], " (",percx,"%",sep=""),
      ylab=paste(colnames(scores)[yy], " (",percy,"%",sep="")
)

contours<-list()
for(g in 1:length(groups)){
  if(is.na(widths[g,1])){
    next
  }
  else {
    contours[[g]]<-contourLines(kde[[g]]$x,
                                kde[[g]]$y,
                                kde[[g]]$z,
                                level=c(quant[[g]][1]))
    for(i in 1:length(contours[[g]])){
      polygon(contours[[g]][[i]]$x,
              contours[[g]][[i]]$y,
              border=NA,
              col=fc[g])
    }
  }
}

```

```

    )
  }
}

for(g in 1:length(groups)){
  if(nrow(groups[[g]])>1){
    points(groups[[g]][,xx],groups[[g]][,yy],
           pch=21,
           col=NA,
           cex=1,
           bg=fc[g])
  }

  if(nrow(groups[[g]])==1){
    text(groups[[g]][,xx],groups[[g]][,yy],
         levels(ts)[g],
         cex=0.5,
         col=fc[g]
        )
  }
  else {
    text(colMeans(groups[[g]])[xx],colMeans(groups[[g]])[yy],
         levels(ts)[g],
         cex=0.5,
         col="black"
        )
  }
}
arrows(0,0,loadings[,xx],loadings[,yy],
       length=0.1,
       angle=15
      )
text(loadings[,xx]*1.2,loadings[,yy]*1.2,rownames(loadings),
     cex=0.7,
     col="darkblue",
     font=2
    )

list(groups=groups, pc=pca)
}

```

pcaplot.clr kde()---unstandardized PCA plot with KDEs

This function generates a PCA plot of unstandardized data with loadings and scores normalized to the eigenvalues. It is designed for handling clr-transformed data, though loading vectors may require alternative scaling. It also generates 100% KDE contours around defined training sets.

Function arguments:

- *x = data*
- *xx = PC axis to plot on the abscissa (1, 2, 3, ...)*
- *yy = PC axis to plot on the ordinate (1, 2, 3, ...)*
- *ts = a vector for defining specimen groups or training sets (members of the same group will be assigned the same colour)*
- *fc = a vector of field colour choices for points and 2D KDE contours for each training set*

Function Values:

- *groups = data clustered by grouping variable*
- *pc = results of the PCA by **prcomp()***

```
pcaplot.clr.kde<-function(x,xx,yy,ts,fc)
{
  pca<-prcomp(x,retx=T,scale=F)
  scores<-array(data=NA,
                dim=c(nrow(pca$x),ncol(pca$x)),
                dimnames=list(rownames(pca$x),colnames(pca$x))
                )
  for (l in 1:ncol(pca$x)){
    scores[,l]<-pca$x[,l]/pca$sdev[l]
  }
  loadings<-array(data=NA,
                  dim=c(nrow(pca$rotation),ncol(pca$rotation)),
                  dimnames=list(rownames(pca$rotation),colnames(pca$rotation))
                  )
  for (m in 1:ncol(pca$rotation)){
    loadings[,m]<-pca$rotation[,m]*pca$sdev[m]
  }
  percx<-round(pca$sdev[xx]^2/sum(pca$sdev^2)*100,2)
  percy<-round(pca$sdev[yy]^2/sum(pca$sdev^2)*100,2)

  groups<-list()
  for(i in 1:nlevels(ts)){
```

```

    groups[[i]]<-subset(scores,ts==levels(ts)[i])
  }
widths<-array(data=NA,c(length(groups),2))
for(j in 1:length(groups)){
  ifelse(nrow(groups[[j]])<4,
    widths[j,1]<-NA,
    widths[j,1]<-width.SJ(groups[[j]][,xx],method="dpi")
  )
  ifelse(nrow(groups[[j]])<4,
    widths[j,2]<-NA,
    widths[j,2]<-width.SJ(groups[[j]][,yy],method="dpi")
  )
}
kde<-list()
for(k in 1:length(groups)){
  if(is.na(widths[k,1])){
    kde[[k]]<-NA
  }
  else {
    kde[[k]]<-kde2d(as.vector(groups[[k]][,xx]),as.vector(groups[[k]][,yy])
,
    h=c(widths[k,1],widths[k,2]),
    n=c(1000,1000),
    lims=c(range(scores[,xx]-5,scores[,xx]+5),
           range(scores[,yy]-5,scores[,yy]+5))
  )
}
}

pkde<-list()
for(p in 1:length(groups)){
  if(is.na(widths[p,1])){
    pkde[[p]]<-NA
  }
  else {
    pkde[[p]]<-pkde2d(groups[[p]][,xx],groups[[p]][,yy],
    h=c(widths[p,1],widths[p,2]),
    n=c(1000,1000),
    lims=c(range(scores[,xx]-5,scores[,xx]+5),
           range(scores[,yy]-5,scores[,yy]+5))
  )
}
}

```

```

}

qe<-seq(0,1,len=10001)

quant<-list()
n_groups<-NULL
for (p in 1:length(groups)){
  if(is.na(widths[p,1])){
    quant[[p]]<-NA
  }
  else {
    n_groups[p]<-nrow(groups[[p]])
    ptdens<-NULL
    for(k in 1:n_groups[p]){
      ptdens<-rbind(ptdens,pkde[[p]]$z[k,k])
    }
    quant[[p]]<-quantile(ptdens,probs=qe)
  }
}

plot(scores[,xx],scores[,yy],
      pch=NA,
      xlim=c(min(scores[,xx],loadings[,xx])-0.5,max(scores[,xx],loadings[,xx]
)+0.5),
      ylim=c(min(scores[,yy],loadings[,yy])-0.5,max(scores[,yy],loadings[,yy]
)+0.5),
      xlab=paste(colnames(scores)[xx],"(",percx,"%",sep=""),
      ylab=paste(colnames(scores)[yy],"(",percy,"%",sep="")
)

contours<-list()
for(g in 1:length(groups)){
  if(is.na(widths[g,1])){
    next
  }
  else {
    contours[[g]]<-contourLines(kde[[g]]$x,
                                kde[[g]]$y,
                                kde[[g]]$z,
                                level=c(quant[[g]][1]))
    for(i in 1:length(contours[[g]])){
      polygon(contours[[g]][[i]]$x,

```

```

        contours[[g]][[i]]$y,
        border=NA,
        col=fc[g]
    )
  }
}
}

for(g in 1:length(groups)){
  if(nrow(groups[[g]])>1){
    points(groups[[g]][,xx],groups[[g]][,yy],
           pch=21,
           col=NA,
           cex=2,
           bg=fc[g])
  }

  if(nrow(groups[[g]])==1){
    text(groups[[g]][,xx],groups[[g]][,yy],
         levels(ts)[g],
         cex=0.8,
         col=fc[g]
        )
  }
  else {
    text(colMeans(groups[[g]])[xx],colMeans(groups[[g]])[yy],
         levels(ts)[g],
         cex=0.7,
         col="black"
        )
  }
}
arrows(0,0,loadings[,xx],loadings[,yy],
       length=0.1,
       angle=15
)
text(loadings[,xx]*1.2,loadings[,yy]*1.2,rownames(loadings),
     cex=0.8,
     col="darkblue",
     font=2
)

```

```
list(groups=groups, pc=pca)
}
```

***ternary()* --- ternary diagram**

This function is an adaptation of the **ternary()** function found in the StatDA package (Filzmoser, 2015). It plots the relative proportions of three component data into a ternary or three pole diagram.

Function arguments:

- *x* = matrix or dataframe with three columns; each row must sum to 100
- *group* = a grouping variable
- *nam* = variable names
- *grid* = a logical argument specifying whether or not to draw a grid
- ... = further graphical parameters

Function values:

- *fc* = a vector of field colours assigned to each group

See **ternary()** in package StatDA (Filzmoser, 2015).

```
ternary<-function (x, group, nam = NULL, grid = FALSE, ...)  
{  
  val = 0.6  
  if (is.null(nam)) {  
    nam <- dimnames(x)[[2]]  
  }  
  s <- rowSums(x)  
  if (any(s <= 0))  
    stop("each row of the input `object' must have a positive sum")  
  dat <- x/s  
  xp <- dat[, 2] + dat[, 3]/2  
  yp <- dat[, 3] * sqrt(3)/2  
  par(pty = "s")  
  
  clusters<-list()  
  for(i in 1:nlevels(group)){  
    clusters[[i]]<-dat[group==levels(group)[i],]  
  }  
}
```

```

fc<-rainbow(length(clusters),s=1,v=1,alpha=0.3)

plot(xp, yp, xlim = c(0, 1), ylim = c(0, 0.9), pch=NA, frame.plot = FALSE,

     xaxt = "n", yaxt = "n", xlab = "", ylab = "", ...)
segments(0, 0, 1, 0)
segments(0, 0, 1/2, sqrt(3)/2)
segments(1/2, sqrt(3)/2, 1, 0)
mtext(nam[1], side = 1, line = -1, at = -0.05, cex = 0.6)
mtext(nam[2], side = 1, line = -1, at = 1.05, cex = 0.6)
text(0.5, 0.9, nam[3], cex = 0.6)

for(g in 1:length(clusters)){
  if(is.null(nrow(clusters[[g]]))==F){
    points(clusters[[g]][,2]+clusters[[g]][,3]/2,clusters[[g]][,3]*sqrt(3)/
2,
          pch=21,
          cex=2,
          col="black",
          bg=fc[g])}
  else{
    points(clusters[[g]][,2]+clusters[[g]][,3]/2,clusters[[g]][,3]*sqrt(3)/
2,
          pch=21,
          cex=2,
          col=NA,
          bg=fc[g])
  }
}

if (grid == TRUE) {
  segments(0.2, 0, 0.1, sqrt(0.03), col = grey(val), lty = "dashed")
  segments(0.4, 0, 0.2, sqrt(0.12), col = grey(val), lty = "dashed")
  segments(0.6, 0, 0.3, sqrt(0.27), col = grey(val), lty = "dashed")
  segments(0.8, 0, 0.4, sqrt(0.48), col = grey(val), lty = "dashed")
  segments(0.2, 0, 0.6, sqrt(0.48), col = grey(val), lty = "dashed")
  segments(0.4, 0, 0.7, sqrt(0.27), col = grey(val), lty = "dashed")
  segments(0.6, 0, 0.8, sqrt(0.12), col = grey(val), lty = "dashed")
  segments(0.8, 0, 0.9, sqrt(0.03), col = grey(val), lty = "dashed")
  segments(0.1, sqrt(0.03), 0.9, sqrt(0.03), col = grey(val),
          lty = "dashed")
}

```

```
segments(0.2, sqrt(0.12), 0.8, sqrt(0.12), col = grey(val),
        lty = "dashed")
segments(0.3, sqrt(0.27), 0.7, sqrt(0.27), col = grey(val),
        lty = "dashed")
segments(0.4, sqrt(0.48), 0.6, sqrt(0.48), col = grey(val),
        lty = "dashed")

text(0.95, 0.21, "20", col = grey(val), cex = 0.6, srt = 60)
text(0.86, 0.35, "40", col = grey(val), cex = 0.6, srt = 60)
text(0.75, 0.54, "60", col = grey(val), cex = 0.6, srt = 60)
text(0.64, 0.72, "80", col = grey(val), cex = 0.6, srt = 60)
text(0.05, 0.21, "80", col = grey(val), cex = 0.6, srt = 300)
text(0.14, 0.35, "60", col = grey(val), cex = 0.6, srt = 300)
text(0.25, 0.54, "40", col = grey(val), cex = 0.6, srt = 300)
text(0.36, 0.72, "20", col = grey(val), cex = 0.6, srt = 300)
text(0.2, -0.02, "20", col = grey(val), cex = 0.6, srt = 60)
text(0.4, -0.02, "40", col = grey(val), cex = 0.6, srt = 60)
text(0.6, -0.02, "60", col = grey(val), cex = 0.6, srt = 60)
text(0.8, -0.02, "80", col = grey(val), cex = 0.6, srt = 60)

}
list(fc=fc)
}
```

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